


Labor_10

SPAM EMAIL CLASSIFICATION

A solid blue horizontal bar spanning the width of the slide, located at the bottom.

processEmail.m

- Lower-casing
 - Stripping HTML
 - Normalizing URLs
 - Normalizing Email Addresses
 - Normalizing Numbers
 - Normalizing Dollars
 - Word Stemming („include“, „includes“, „included“ -> „includ“)
 - Removal of non-words
- 

preprocessEmail.m

```
anyon know how much it cost to host a web portal well it depend on how  
mani visitor your expect thi can be anywher from less than number buck  
a month to a coupl of dollarnumb you should checkout httpaddr or perhap  
amazon ecnumb if your run someth big to unsubscrib yourself from thi  
mail list send an email to emailaddr
```

Figure 9: Preprocessed Sample Email

```
1 aa  
2 ab  
3 abil  
...  
86 anyon  
...  
916 know  
...  
1898 zero  
1899 zip
```

Figure 10: Vocabulary List

```
86 916 794 1077 883  
370 1699 790 1822  
1831 883 431 1171  
794 1002 1893 1364  
592 1676 238 162 89  
688 945 1663 1120  
1062 1699 375 1162  
479 1893 1510 799  
1182 1237 810 1895  
1440 1547 181 1699  
1758 1896 688 1676  
992 961 1477 71 530  
1699 531
```

Figure 11: Word Indices for Sample Email

Vocabulary List

The next step is to choose which words we would like to use in our classifier and which we would want to leave out.

Convert each email into a vector. Specifically whether the i -th word in the dictionary occurs in the email.:

$$x = \begin{bmatrix} 0 \\ \vdots \\ 1 \\ 0 \\ \vdots \\ 1 \\ 0 \\ \vdots \\ 0 \end{bmatrix} \in \mathbb{R}^n$$